

Here ' b ' denote sites. There are ' B ' sites in total. ' d ' denotes detector tube and there are ' D ' tubes in total. $P(b, d)$ is the probability that emission in ' b ' is detected in tube ' d '. The emission at site ' b ' is assumed to be Poisson with mean $\lambda(b)$.

$n^*(d)$ is the total number of emissions detected in tube d during acquisition (real data).

$n(b, d)$ denotes the number of emissions in ' b ' detected in ' d '. $n(b, d)$ are mutually independent Poisson random variable, across ' b ', across ' d ' (split of Poisson process independent).

$$n(b, d) \sim \text{Poisson}(\lambda(b, d))$$

$$\lambda(b, d) = \lambda(b)p(b, d)$$

$$\sum_{d=1}^D p(b, d) = 1$$

$$n^*(d) = \sum_{b=1}^B n(b, d)$$

$$\lambda^*(b, d) = E[n^*(d)] = \sum_{b=1}^B \lambda(b)p(b, d)$$

Best estimate of $\lambda^*(d)$ is $n^*(d)$

$$n^*(d) = \sum_{b=1}^B \lambda(b)p(b, d) \quad \text{for } d = 1, 2, \dots, D$$

Above sets of linear equations form the basis for inversion reconstruction.

Now let's move onto expectation maximization. If we have $n(b, d)$ as hidden random variable then estimation of $\lambda(b)$ becomes very easy.

So,

$$X = \text{hidden variable} = n(b, d) = x$$

$$Y = \text{observed data} = p^*(d) = y$$

$$\text{Likelihood} = p(x, y) = \prod_{\substack{b=1, 2, \dots, B \\ d=1, 2, \dots, D}} \frac{e^{-\lambda(b, d)} (\lambda(b, d))^{n(b, d)}}{n(b, d)!}$$

Each $n(b, d)$ poisson process is independent

$$\sum_{b=1}^B n(b, d) = n^*(d) \quad \text{for all } 'd'$$

$$\log(\text{Likelihood}) = \sum_{d=1}^D \sum_{b=1}^B -\lambda(b, d) + n(b, d) \log(\lambda(b, d)) - \log(n(b, d)!)$$

$\log(n(b, d)!)$ of no use since it doesn't contain parameter $\lambda(b, d)$

$$E[n(b, d) | n^*(d)] = \frac{n^*(d)\lambda^i(b, d)}{\sum_{b=1}^B \lambda^i(b, d)} \quad \lambda^i(b, d) = \text{estimate at iteration } i$$

$$E[\log(\text{Likelihood})] = \sum_{d=1}^D \sum_{b=1}^B -\lambda(b, d) + \frac{n^*(d)\lambda^i(b, d)\log(\lambda(b, d))}{\sum_{b=1}^B \lambda^i(b, d)} = Q(\theta, \theta^i)$$

Differentiating above expression respect to $\lambda(b, d)$ to get maxima

$$\lambda^{i+1} = \frac{n^*(d)\lambda^i(b, d)}{\sum_{b'=1}^B \lambda^i(b', d)}$$

$$\lambda^i(b, d) = \lambda^i(b)p(b, d) \quad \text{and} \quad \lambda(b) = \sum_{d=1}^D \lambda(b, d)$$

$$\lambda^{new}(b) = \lambda^{old}(b) \sum_{d=1}^D \frac{n^*(d)p(b, d)}{\sum_{b'=1}^B \lambda^{old}(b')p(b', d)}$$

Final formula for update is given below

$$\lambda^{new}(b) = \lambda^{old}(b) \sum_{d=1}^D \frac{n^*(d)p(b, d)}{\sum_{b'=1}^B \lambda^{old}(b')p(b', d)}$$